

Last lecture

Random Variables (RV)

- Probability Mass Function (pmf)
- Mean and Variance (Ch 2.2)

Agenda

Random Variables (RV)

- Mean and Variance (Ch 2.2)

Conditional Probability (Ch 2.3)

- Motivation
- Examples
- Solver

Law of Total Probability (Ch 2.10)

Variance and Standard Deviation

Mean is important... but not complete enough

- Variance $Var(X)$ is how PMF spreads apart from μ_X

- $Var(X) \triangleq E[(X - \mu_x)^2] = E[X^2] - (\underbrace{E[X]}_{\mu_x})^2$

\downarrow
 $E[X]$

unit². e.g. m² if $x = ? m$.

Variance and Standard Deviation

Standard deviation $\sigma_X \triangleq \sqrt{\text{Var}(X)}$; $\text{Var}(X) = \sigma_X^2$

exam-safe

σ_X is of the same unit as X

$$\text{Var}(X + c) = E[(X + c) - E[X + c]]^2 = E[(X - E[X]) + c - E[c]]^2 = E[(X - E[X]) + 0]^2 = E[(X - E[X])^2] = \text{Var}(X)$$

$$\text{Var}(aX + c) =$$

2X vs X + X

$$E[(aX)^2] - (E[aX])^2 = \text{Var}(X)$$

$$\text{Var}(aX) = a^2 \text{Var}(X) = E[a^2 X^2] - (a \mu_X)^2 = a^2 (E[X^2] - \mu_X^2) = a^2 \text{Var}(X)$$



$$Y = X + c \quad \text{Var}(Y) = E[(Y - \underbrace{E[Y]}_{E[X+c] = \underline{E[X]} + c})^2]$$

$$= E[(X + \cancel{c} - (E[X] + \cancel{c}))^2]$$

$$\text{Var}(2X) = 2^2 \text{Var}(X) = 4 \text{Var}(X)$$

$$\begin{aligned} \text{Var}(X_1 + X_2) &= E[(X_1 + X_2 - \mu_{X_1} - \mu_{X_2})^2] \\ &= E[(X_1 - \mu_{X_1}) + (X_2 - \mu_{X_2})]^2 \\ &= \text{Var}(X_1) + \text{Var}(X_2) = 2 \text{Var}(X) \end{aligned}$$

Standardized RV

For any RV X

- $\frac{X - \mu_X}{\sigma_X}$ is a standardized RV – Mean 0, variance 1



$$E\left[\frac{X - \mu_X}{\sigma_X}\right] = \frac{E[X] - \mu_X}{\sigma_X} = \frac{\mu_X - \mu_X}{\sigma_X} = 0$$

$$\text{Var}\left[\frac{X - \mu_X}{\sigma_X}\right] = \left(\frac{1}{\sigma_X}\right)^2 \text{Var} X = \frac{\sigma_X^2}{\sigma_X^2} = 1$$

Mean μ_x $E[f(x)] = f(E[x])$

$$\begin{cases} E[ax+c] = aE[x] + c \\ \text{Var}(aX+c) = a^2 \text{Var}(X) \end{cases}$$

$\Rightarrow \text{Var}(X+Y) = \text{Var } X + \text{Var } Y$
if X, Y are independent.

Conditional Probability

Motivation

- The probability of B happens given A happens
- $P(\text{pair of socks are same color})$ given $S_1 = B$
 - $P(I \text{ win Texas Hold'em})$ given $X = \text{Ace} + \text{Ace}$
 - $P(I \text{ pass 313})$ given I skip HW1...

Why do we need conditional probability?

- Analyze the relationship between two events
- Find the optimal solution to make an event probable

Examples

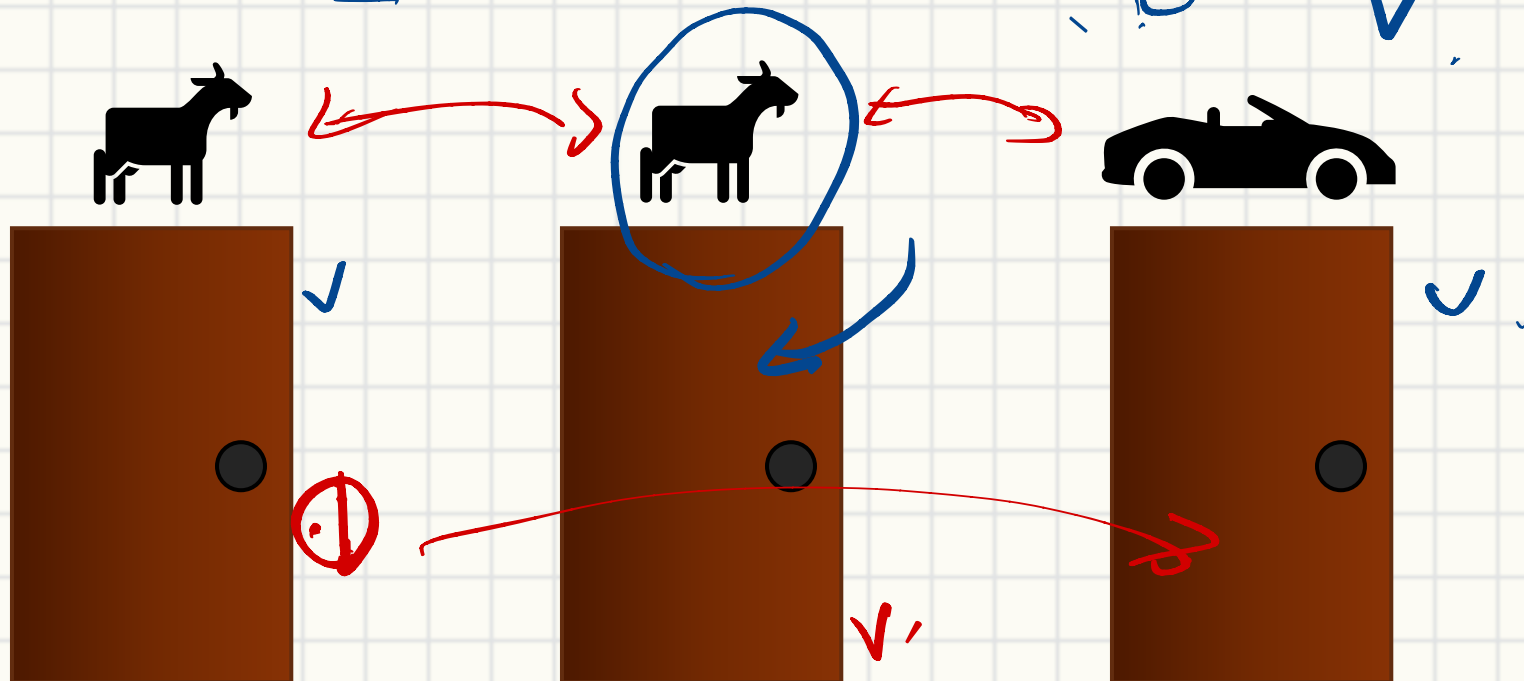
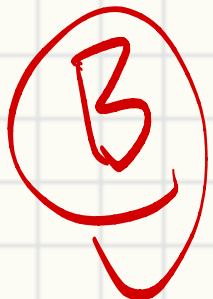
$$P(\text{win} | x_i) \rightarrow$$

① 33%

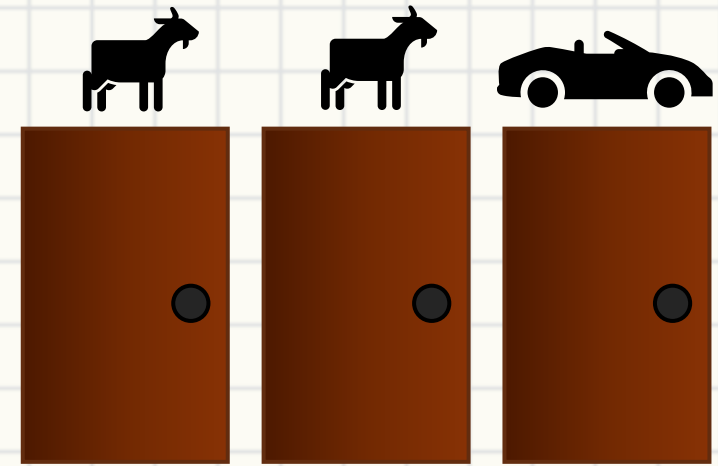
50. (2) 50% 67%

3 doors problem

- 3 closed doors – 1 leads to a car, the others lead to goats
- After you choose one, the host will open a “Fail” door
- Should you change the door?



Examples



- Never change

- $P(W|X_1 = C) = 100\%$
- $P(W|X_1 = G) = 0\%$

$$P(X_1 = C)$$

$$\begin{cases} P(W, X_1 = C) = 100\% \times \frac{1}{3} = 33\% \\ P(W, X_1 = G) = 0\% \end{cases}$$

- Change

- $P(W|X_1 = C) = 0\%$
- $P(W|X_1 = G) = 100\%$

$$\begin{cases} P(W, X_1 = C) = 0\% \\ P(W, X_1 = G) = 100\% \times \frac{2}{3} = 67\% \end{cases}$$

- What if there are 4 doors... 2 cars and 2 goats?

2C 2G

No changing. 50%

Change. $\left\{ \begin{array}{l} P(W | X_1 = G) = 100\% \end{array} \right.$

$\left\{ \begin{array}{l} P(W | X_1 = C) = 50\% \end{array} \right.$ $P(X_1 = C)$

75% + $P(W, X_1 = G) = 1 \times \frac{1}{2} = \frac{1}{2}$
 $P(W, X_1 = C) = 50\% \times \frac{1}{2} = 25\%$

Conditional Probability

$$\underline{P(B|A)} = \begin{cases} \frac{P(A \cap B)}{P(A)} & \text{if } P(A) > 0 \\ \text{undefined} & \text{if } P(A) = 0 \end{cases}$$

Roll two dice, A = sum is 6; B = numbers are not equal

$$P(B) = ? \quad P(B|A) = ? \quad P(B^c|A) = ?$$